

More Than Relevance: High Utility Query Recommendation By Mining Users' Search Behaviors



Xiaofei Zhu, Jiafeng Guo, Xueqi Cheng, Yanyan Lan Institute of Computing Technology, Chinese Academy of Sciences zhuxiaofei@software.ict.ac.cn, {guojiafeng, cxq, lanyanyan}@ict.ac.cn



2. KEY IDEA how to infer query utility? Key Idea: Through user's search behaviors A typical search session



3. QUERY UTILITY MODEL



 $P(C_i = 1 | R_i = 1, A_i = 1) = 1,$ $P(A_i = 1) = \alpha_{\phi(i)},$ $P(S_i = 1 | C_{1:i}) = \sigma(\sum \beta_{\phi(k)} \cdot I(C_k = 1)),$ $\sigma(x) = \frac{1}{1+e^{-x}}$ $P(R_i = 1 | R_{i-1} = 1, S_{i-1} = 1) = 0.$

4. EVALUATION METRICS Original query our Original query previous **Recommendations & Search Results** Recommendations query 1 **Relevant or Not?** query 1 **Relevant or Not?** doc 1 doc 3 doc 2 query 2 **Relevant or Not?** query 2 doc 1 doc 3 doc 2 **Relevant or Not?** query 3

Perceived Utility α : control the probability of the attractiveness

Posterior Utility β : control the probability of users' satisfaction

- R_i: whether there is a reformulation at position i
- C_i : whether the user clicks on some of the search results of the reformulation at position i;
- : whether the user is attracted by the search results of the reformulation at position I;
- S_i : whether the user's information needs have been satisfied at position i;

Query Utility $\mu_t = \alpha_t * \beta_t$

The expected information gain users obtained from the search results of the query according to their original information needs

query 3 Relevant or Not?



– QRR (Query Relevant Ratio) $QRR(q) = \frac{RQ(q)}{q}$ N(q)

Measuring the probability that a user finds(clicks) relevant results when she uses query q for her search task.

– MRD (Mean Relevant Document) $MRD(q) = \frac{RD(q)}{N(q)}$

Measuring the average number of relevant results a user finds(clicks) when she uses query q for her search task.

5. EXPERIMENTAL RESULTS



6. CONCLUSIONS

Contribution

- Recommend high utility queries rather than only relevant queries: to directly toward the ultimate goal of query recommendation;
- A novel dynamic Bayesian network (i.e., QUM) to mine query utility from users' reformulation and click behaviors;

| (a) QRR | | | (b) MRD | | |
|------------------|---------------|---------------|---------------|---------------|---------------|
| Query Difficulty | Method | QRR | | MRD | |
| | | @5 | @10 | @5 | @10 |
| Easy | ADJ | 0.588(18.64%) | 0.526(26.30%) | 0.771(20.32%) | 0.674(25.22%) |
| | CO | 0.609(14.55%) | 0.529(25.63%) | 0.830(11.80%) | 0.687(22.89%) |
| | \mathbf{QF} | 0.618(12.94%) | 0.604(9.89%) | 0.846(9.67%) | 0.806(4.69%) |
| | \mathbf{CT} | 0.654(6.62%) | 0.635(4.65%) | 0.836(11.02%) | 0.805(4.79%) |
| | PCU | 0.656(6.37%) | 0.611(8.74%) | 0.889(4.35%) | 0.798(5.79%) |
| | PTU | 0.689(1.22%) | 0.663(0.17%) | 0.908(2.18%) | 0.837(0.86%) |
| | QUM | 0.698 | 0.664 | 0.928 | 0.844 |
| Medium | ADJ | 0.460(30.00%) | 0.429(33.19%) | 0.596(24.14%) | 0.527(33.76%) |
| | CO | 0.495(20.81%) | 0.441(29.65%) | 0.640(15.72%) | 0.550(28.10%) |
| | \mathbf{QF} | 0.511(17.07%) | 0.500(14.39%) | 0.615(20.43%) | 0.630(11.79%) |
| | \mathbf{CT} | 0.534(12.07%) | 0.549(4.02%) | 0.689(7.54%) | 0.692(1.81%) |
| | PCU | 0.544(9.91%) | 0.485(17.74%) | 0.703(5.31%) | 0.588(19.76%) |
| | PTU | 0.581(2.87%) | 0.557(2.70%) | 0.722(2.53%) | 0.689(2.18%) |
| | QUM | 0.598 | 0.572 | 0.740 | 0.704 |
| Hard | ADJ | 0.259(65.27%) | 0.216(91.19%) | 0.351(54.37%) | 0.284(77.27%) |
| | CO | 0.314(36.29%) | 0.261(58.17%) | 0.412(31.63%) | 0.340(48.00%) |
| | \mathbf{QF} | 0.324(32.08%) | 0.312(32.20%) | 0.441(22.94%) | 0.414(21.78%) |
| | \mathbf{CT} | 0.334(28.08%) | 0.343(20.17%) | 0.437(24.15%) | 0.424(18.85%) |
| | PCU | 0.404(5.90%) | 0.324(27.07%) | 0.534(1.54%) | 0.413(22.02%) |
| | PTU | 0.426(0.28%) | 0.402(2.51%) | 0.526(3.18%) | 0.485(3.92%) |
| | QUM | 0.427 | 0.412 | 0.542 | 0.504 |

The performance improvements are significant (t-test, p-value <= 0.05)

- Introduce two evaluation metrics for utility based recommendation
- Evaluate the performance on a real query log and show the effectiveness

Future work

- Extend our utility model to capture the specific clicked URLs for finer modeling
- Extend our utility model to capture the query level utility